





Achieving Reliability in Master-Worker Computing via Evolutionary Dynamics

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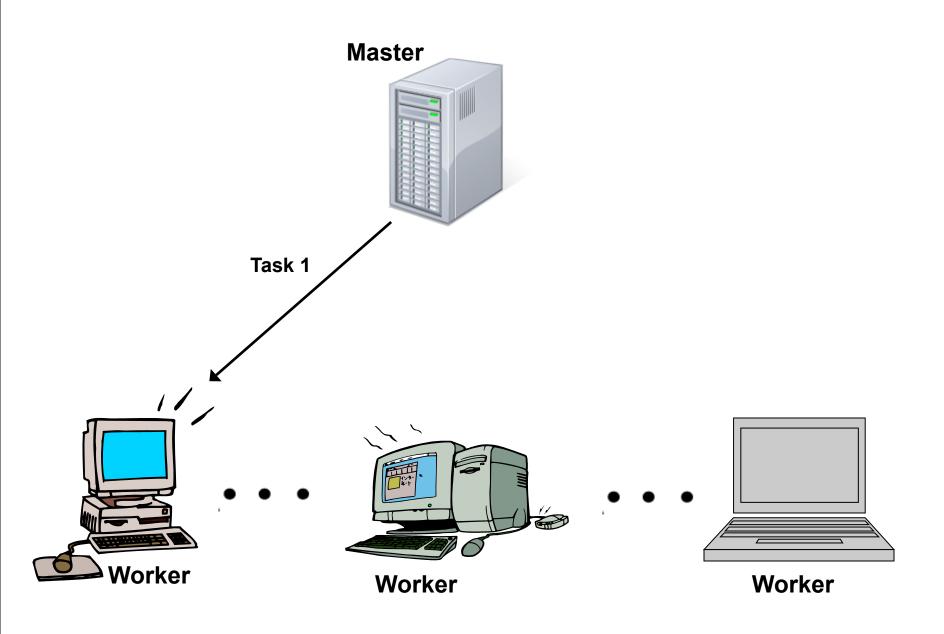
July 17th, 2012

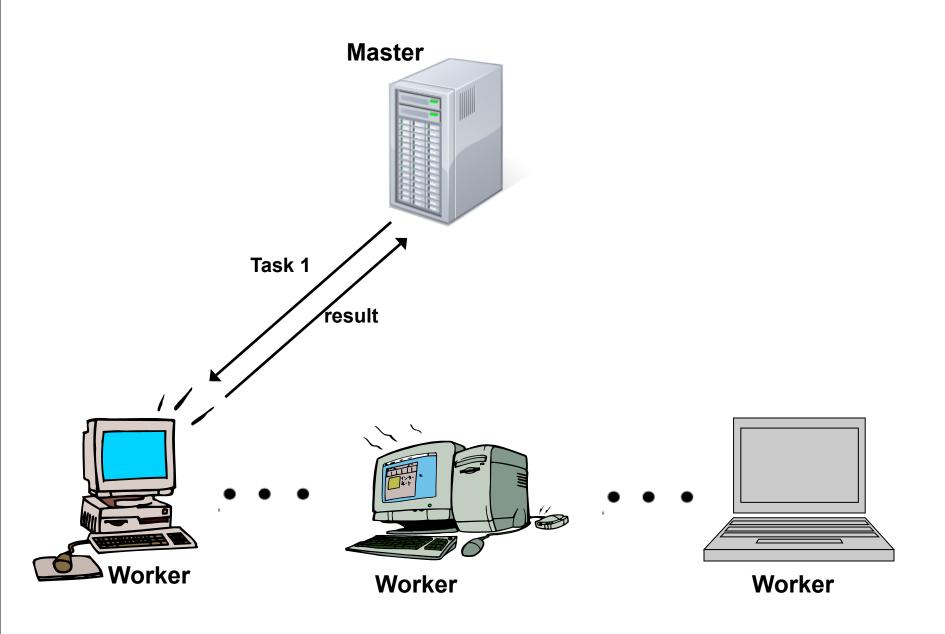
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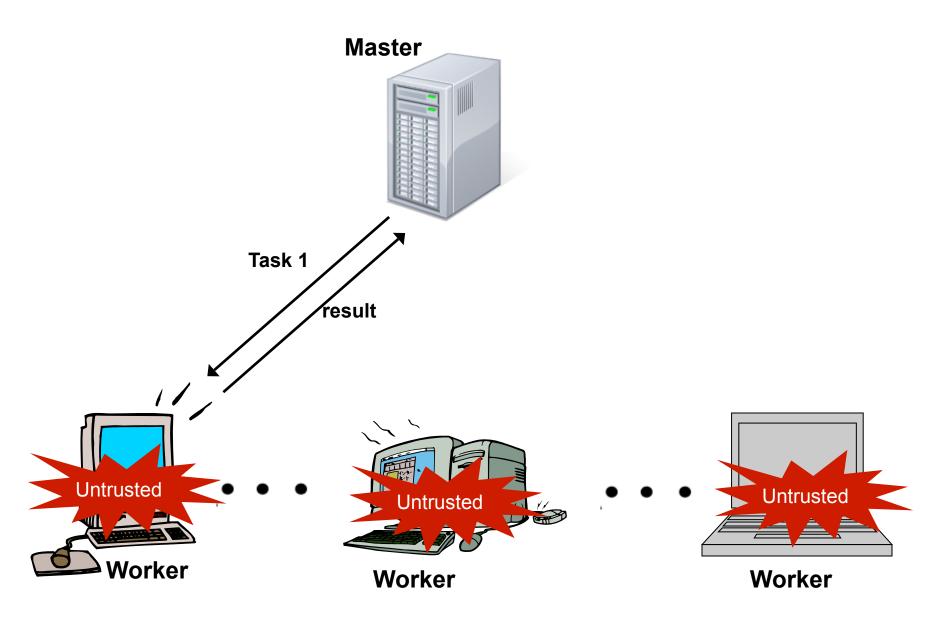
This work is supported in part by the Cyprus Research Promotion Foundation grant TΠΕ/ΠΛΗΡΟ/0609(BE)/05

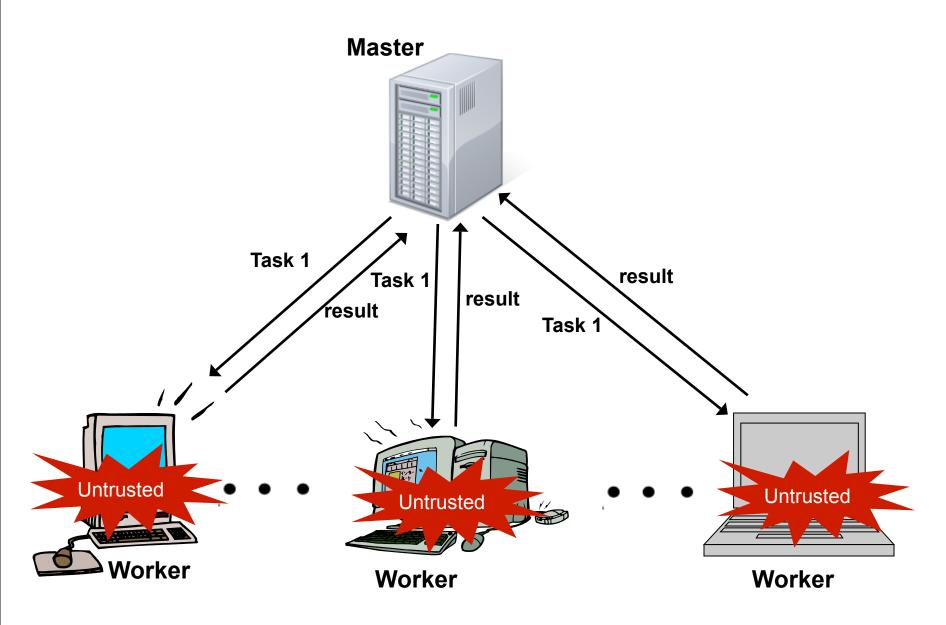


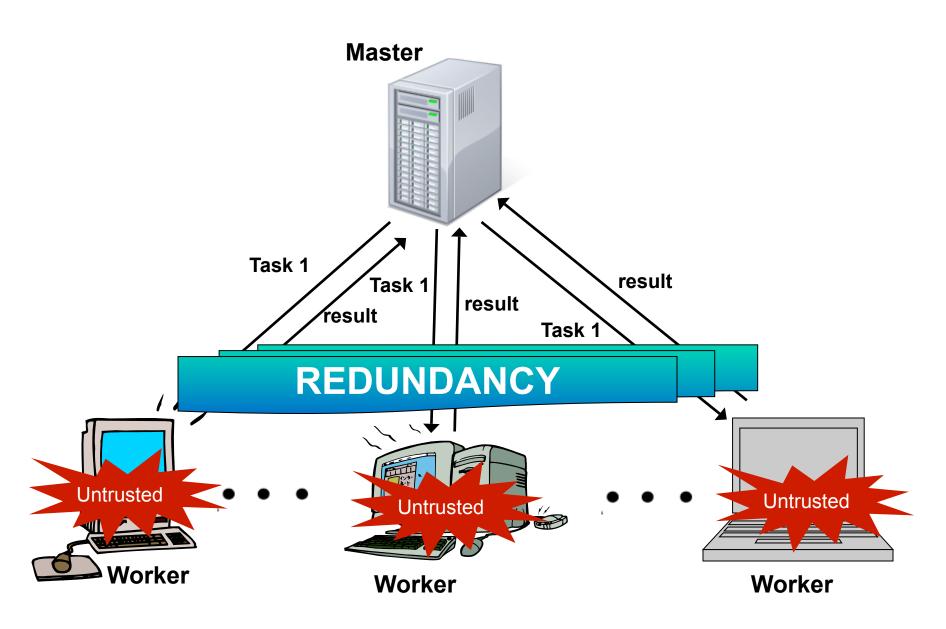


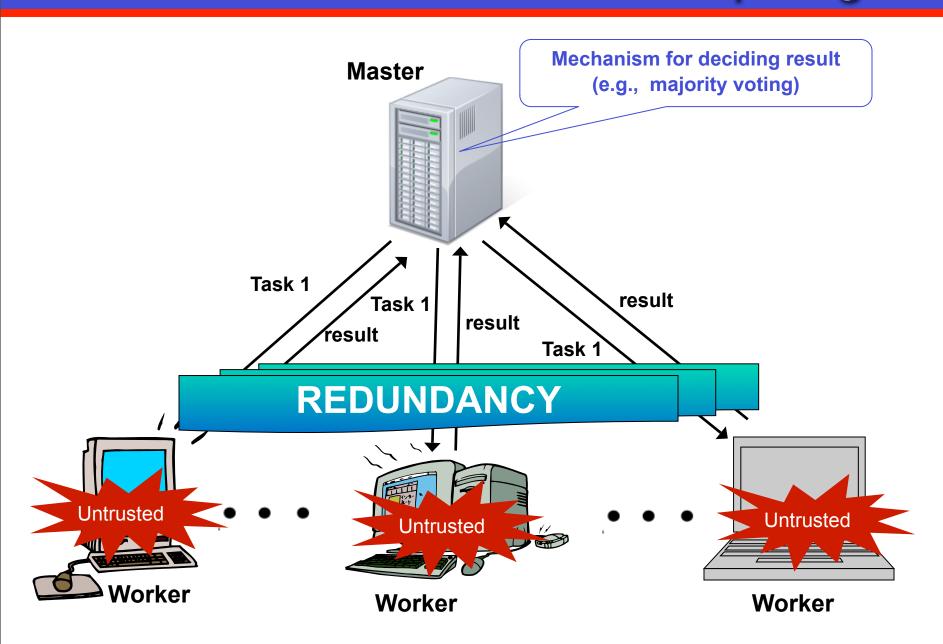












- Rational workers: act upon their best interest, i.e., choose the strategy that maximizes their own benefit
 - Honest: compute and report correct result
 - Cheat: fabricate and return a bogus result
- Mechanisms with reward/punish schemes that provide incentives to workers to be honest
 - One shot: in each round a task is performed and no knowledge is forwarded to the next round

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[Shneidman Parkers 03]

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- Employ reinforcement learning both on Master and Workers
 - Positive payoffs increase probability of strategy just chosen
 - Negative payoffs reduce the probability
 - □ Knowledge only of the payoffs received, not of the strategies involved [Camerer 03,Szepesvari 10]

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Objective: Develop a reliable computation platform where the master obtains the correct task results (whp).

Model: Master

 $p_{\mathcal{A}}$

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- The Master can audit the responses (with some cost)
 - Auditing means performing the task
 - $\ \ \ \ p_{\mathcal{A}}$:probability of auditing
 - > It may change from round to round
- Eventual correctness: After some finite number of rounds, the master obtains the correct task in every round, with minimal auditing, while keeping the workers satisfied

Model: Workers



Model: Workers

- Each worker i has
 - $extstyle extstyle extstyle extstyle a probability of cheating <math>p_{C_i}$
 - > It may change from round to round
 - \Box an aspiration a_i
 - > the minimum benefit it expects to obtain in a round
- Payofffs

WP_C	Worker's punishment for being caught cheating
WC_T	Worker's cost for computing a task
WB_{y}	Worker's benefit from master's acceptance

Master's Protocol

Set initial p_A (e.g., 0.5)

Repeat

Send a task to all *n* workers

Upon receiving all answers do

Audit the answers with probability p_A

If the answers were not audited then

Accept the value returned by the majority

Else

$$p_{\mathcal{A}} \leftarrow p_{\mathcal{A}} + \alpha_m \cdot \left(\frac{cheaters}{n} - \tau\right)$$

Give appropriate payoff Π_i to each worker i

 α_m : learning rate (tunes the extent of change)

 τ : tolerance (tolerable ration of cheaters, e.g., 0.5)

Protocol for Worker i

Set initial p_{C_i} (e.g., 0.5)

Repeat

Receive a task from the master

Set $S_i = -1$ with probability p_{C_i} , $S_i = 1$ otherwise

If $S_i = 1$ then compute the task and send the result

Else send an arbitrary result

Get payoff Π_i

$$p_{C_i} \leftarrow p_{C_i} - \alpha_w \cdot (\Pi_i - a_i) \cdot S_i$$

 α_w : learning rate (tunes the extent of change)

• We analyze the evolution of the master-worker system as a Markov chain and we show:

- Convergence time: The number of rounds to achieve eventual correctness
 - We show, both in expectation and with high probability, that our mechanism reaches convergence time quickly
 - We complement the analysis with simulations.

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$$WB_y \ge a_i + WC_T, \ \forall i \in \mathbb{Z}, \ |\mathbb{Z}| > n/2$$

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Examples of Convergence

 Under certain conditions, the expected convergence time is

$$\left(\alpha_w \cdot (WB_y - WC_T - \max_i \{a_i\}) \cdot \varepsilon\right)^{-1}$$

where

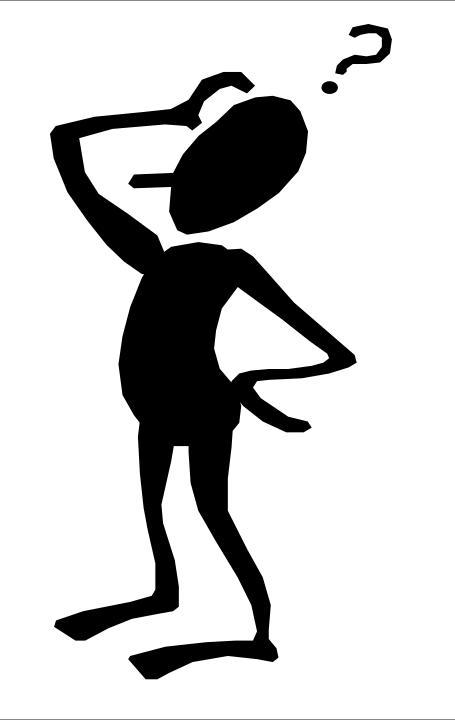
$$\varepsilon \in (0, 1 - (WC_T + \max_i \{a_i\}) / WB_y).$$

Under certain conditions, the converge time is at most

with probability at reast $\frac{\ln(1/\varepsilon)/p + 1/dec}{\ln(1/\varepsilon)/p}$

where
$$(1-arepsilon)(1-e^{-n/96})(1-e^{-n/36})^{1/dec}$$

$$dec = \min_{i} \{\alpha_w \cdot \min\{a_i, WB_y - WC_T - a_i\}\}, \text{ and } \varepsilon \in (0, 1)$$



Thank you!

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Internet-based Task Computing

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- Internet emerges as a viable platform for supercomputing
 - Grid and Cloud computing
 - > e.g., EGEE Grid, TERA Grid, Amazon's EC2
 - Master-Worker volunteer computing: @home projects
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Great potential limited by untrustworthy entities



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 - □ CPUs: 88,128 SPARC64 VIIIfx 8-core 2.0GHz
 - □ 11,280 TFLOPS (11.2 PetaFLOPS)

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http://setiathome.berkeley.edu/

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Comparable processing power with top Supercomputers

@ a fraction of the cost!